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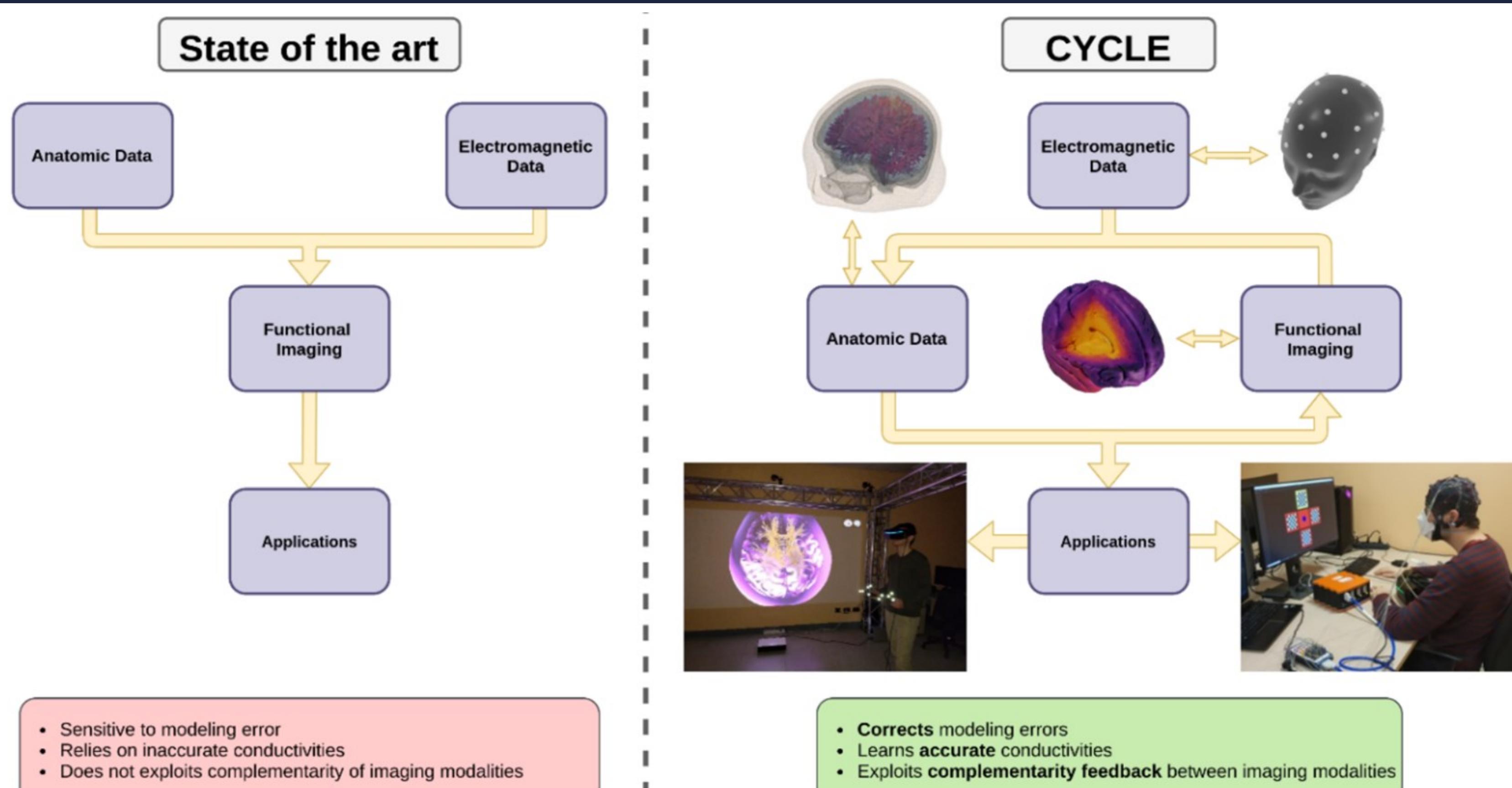
D. Ben Salem

**IETR**

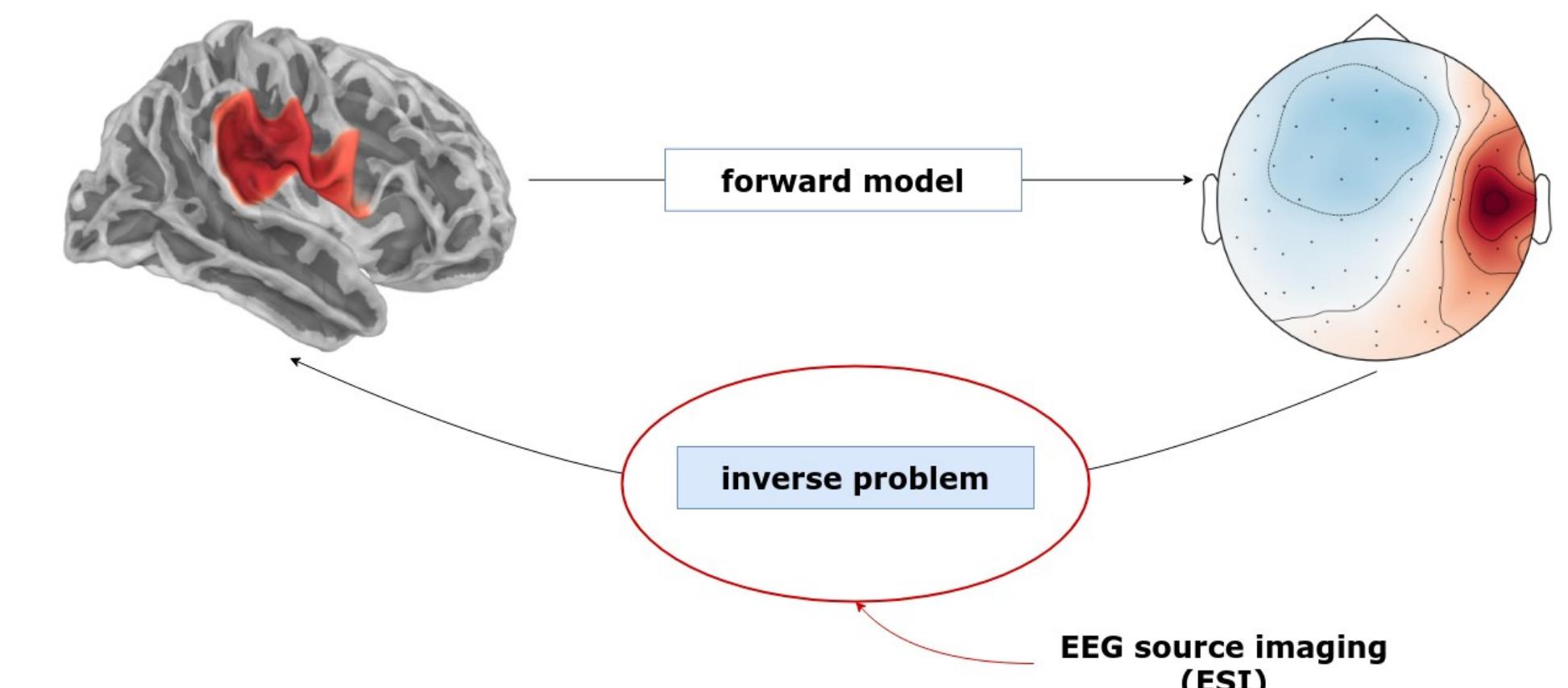
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## General description



## Inverse problem in EEG



$$\text{Forward model: } \mathbf{Y} = \mathbf{LX} + \boldsymbol{\varepsilon} \quad (\boldsymbol{\varepsilon} = \text{additive noise})$$

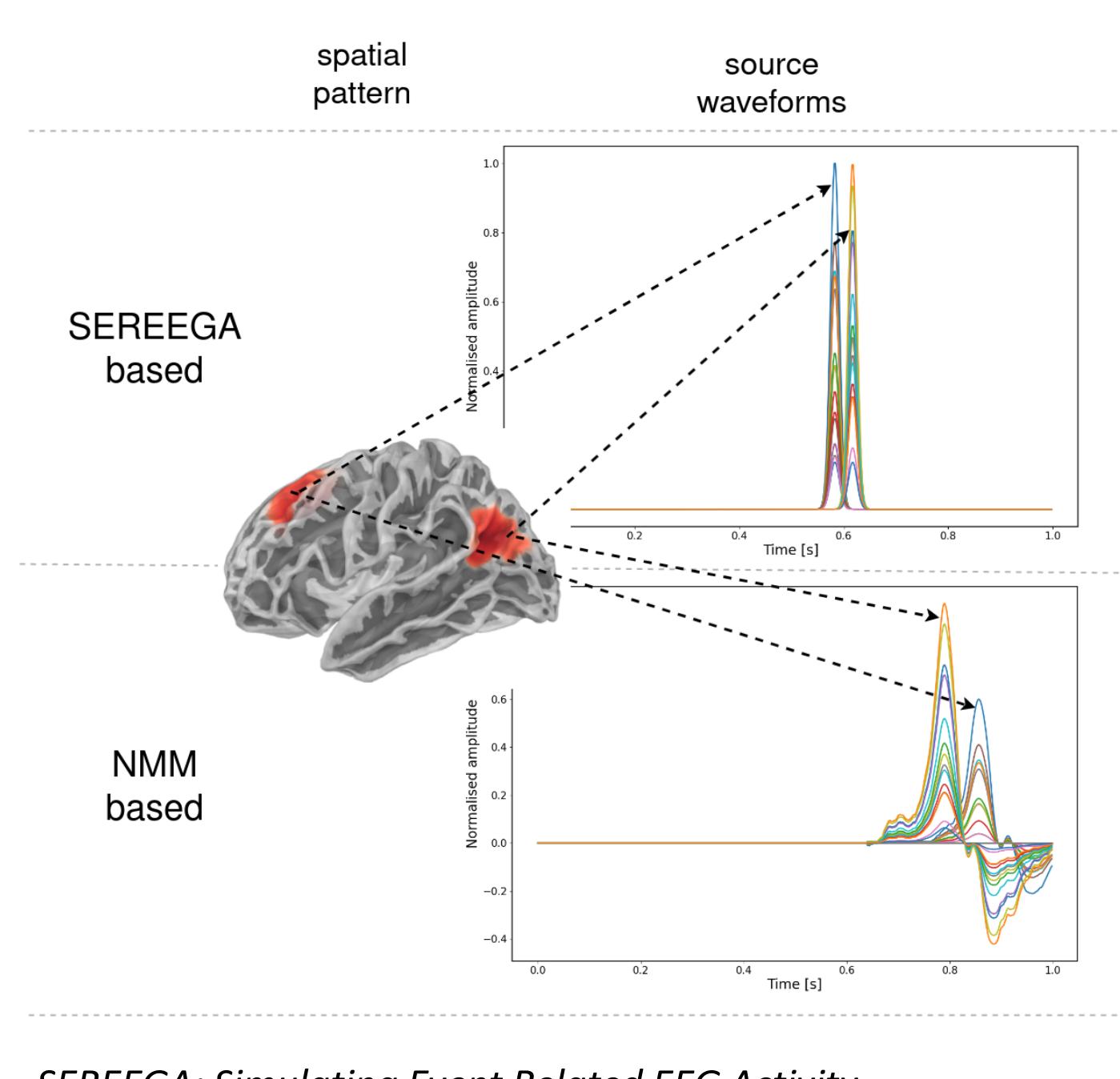
- !  $N_s \gg N_e$ :  $\mathbf{L}$  non-invertible
- Volume conduction/mixing

III-posed inverse problem: add prior on  $\mathbf{X}$  to solve

$$\hat{\mathbf{X}} = \underset{\mathbf{X}}{\operatorname{argmin}} \underbrace{\|\mathbf{Y} - \mathbf{LX}\|_F^2}_{\text{data fitting}} + \underbrace{\lambda R(\mathbf{X})}_{\text{prior}}$$

## Supervised learning for ESI

### Data simulation



### Direct inversion\*

3 NNs used: 1D CNN, LSTM, spatio-temporal (ResNet + LSTM)  
2 simulated dataset + real data: evaluate generalization ability

train - test	GT	1DCNN	LSTM	deepSIF	MNE	sLORETA
SEREEGA - SEREEGA						
Loc. Error [mm]	0	2.99	5.17	<b>2.64</b>	14.51	9.17
SEREEGA - nmm						
Loc. Error [mm]	0	<b>7.38</b>	9.39	11.86	15.65	12.12
Real data						

\*Accepted in *Frontiers in Neuroscience* [3]

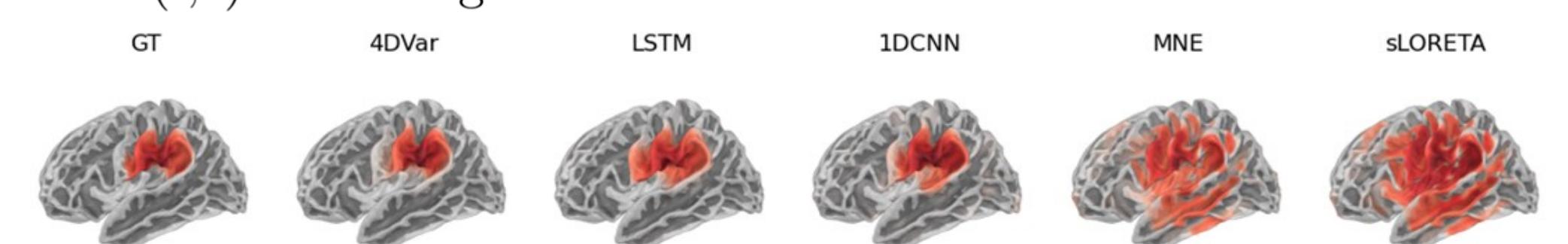
### Model based

Bi-level optimisation problem

$$\theta_\phi^*, \theta_\psi^* = \underset{\theta_\phi, \theta_\psi}{\operatorname{argmin}} \mathcal{L}(\mathbf{X}, \hat{\mathbf{X}}) \quad (O)$$

$$\text{s.t. } \hat{\mathbf{X}} = \underset{\mathbf{X}}{\operatorname{argmin}} f_o(\mathbf{Y}, \mathbf{LX}) + \lambda f_p(\mathbf{X}, \phi(\mathbf{X})) \quad (I)$$

- $\theta_\phi, \theta_\psi$  : parameters of  $\phi$  and  $\psi$  NNs
- $f_o, f_p$  : discrepancy measures (e.g.  $f_o(\mathbf{A}, \mathbf{B}) = \|\mathbf{A} - \mathbf{B}\|_F^2$ )
- $\mathcal{L}(., .)$  : training loss function

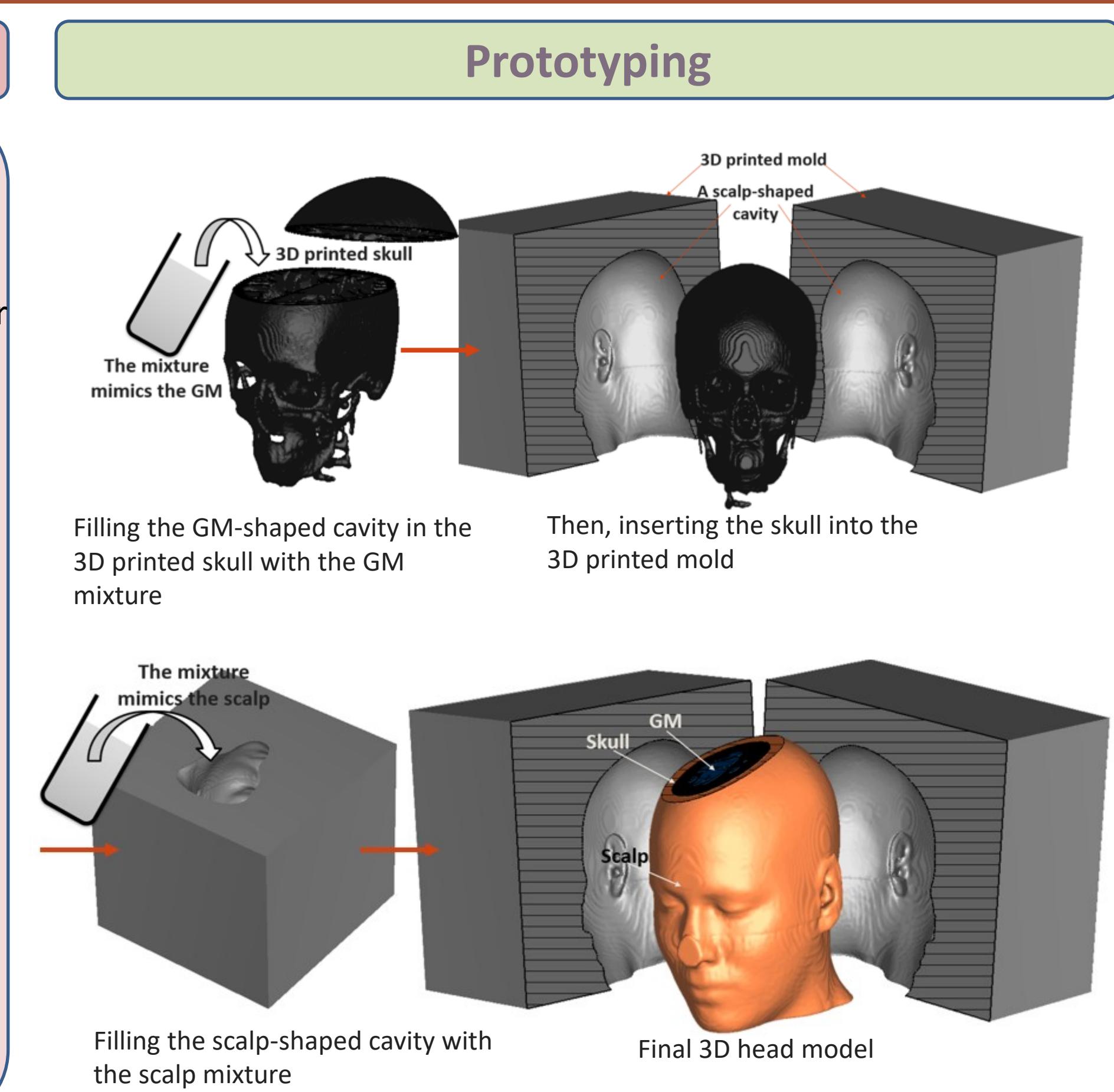
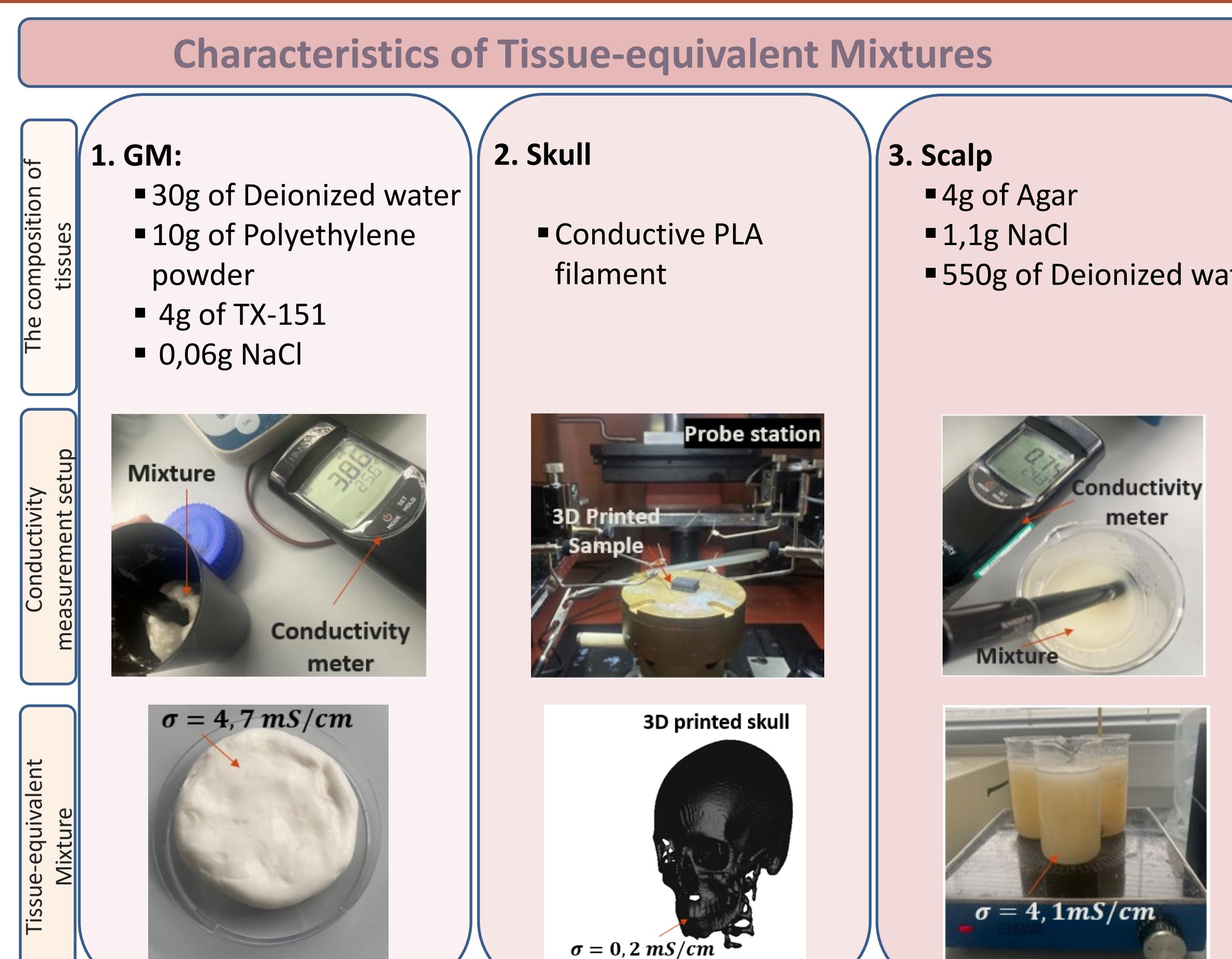
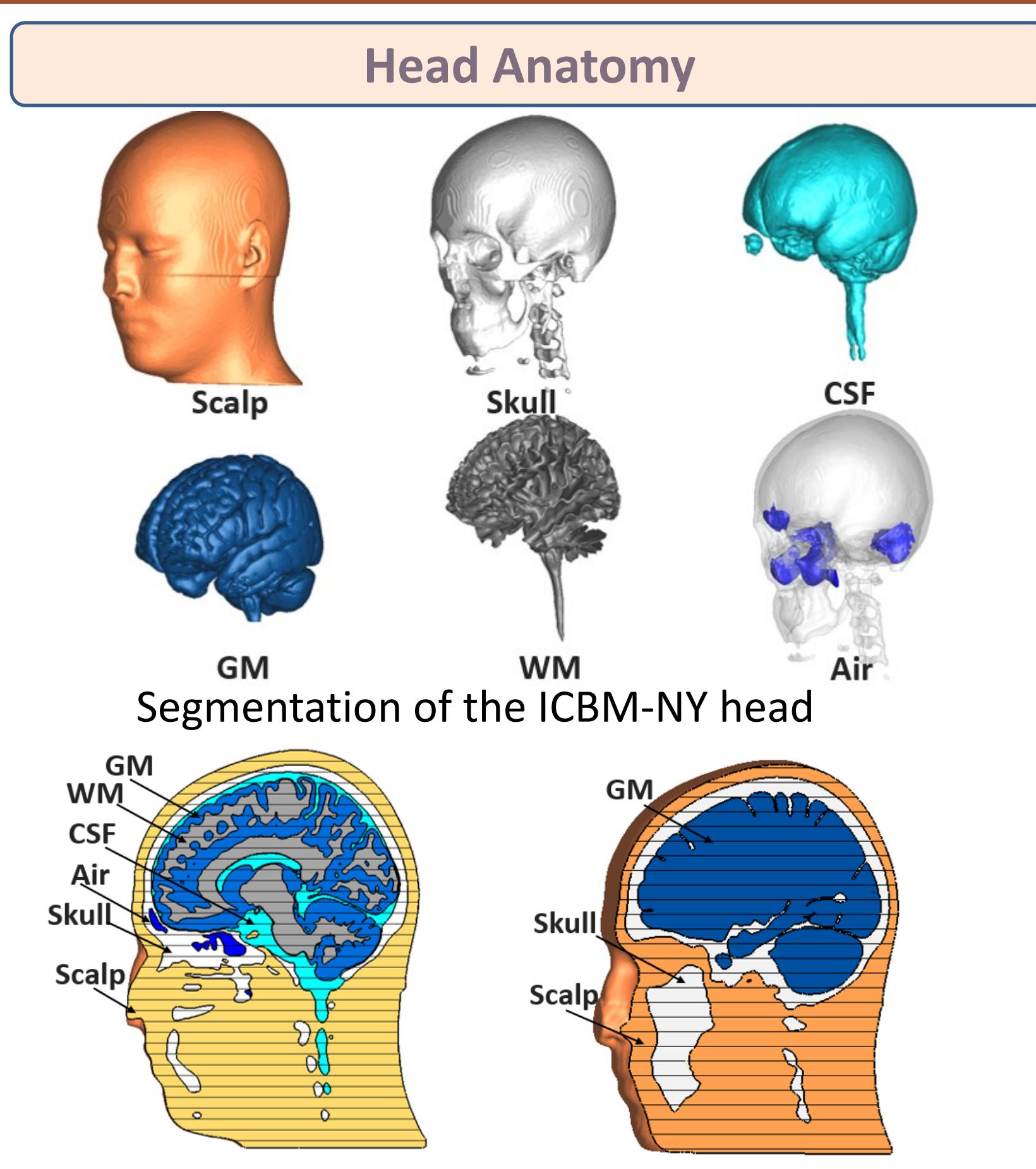


First results: performances close to direct inversion

Future directions:

- Many open questions → work to improve results
- Whole evaluation, test on real data

## Electromagnetic tissue-equivalent phantoms for functional brain imaging



## Publications

- V. Giunzioni et al., *IEEE Trans. Antennas Propagat.*, submit.
- R Chen et al., *IEEE Trans. Antennas Propagat.*, in rev.
- S. Reynaud et al., *Frontiers in Neuroscience*, 2024.
- R. Chen et al., ACES-China, 2024.
- A. Scazzola et al., AP-S/INC-USNC-URSI, 2024
- S. Reynaud et al., EUSIPCO 2023.
- A. Merlini et al., EMTS 2023.
- A. Merlini et al., *IEEE Trans. Antennas Propagat.*, 2022.
- A. Merlini et al., *IEEE Trans. Antennas Propagat.*, 2023.
- A. Merlini et al., ICEAA 2023.
- C. Henry et al., IEEE CAMA 2023.
- C. Henry et al., EuCAP 2023.
- V. Giunzioni et al., *Journal of Comput. Physics*, 491, 2023.
- V. Giunzioni et al., IEEE CAMA 2023.
- C. Henry et al., *IEEE Trans. Antennas Propagat.*, 2022.
- C. Henry et al., AP-S/USNC-URSI 2022.
- C. Henry et al., CAMA 2022.
- D. Consoli et al., ICEAA 2022.
- C. Henry et al., URSI GASS 2021.
- S. B. Adrian et al., *IEEE Open J. Antennas Propag.*, 2, 2021.